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Magnetically controllable attitude trajectory constructed using the particle swarm optimization method

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Introduction

Problem:

ensuring spacecraft 3-axis stabilization using magnetic attitude control system only

Restriction:

direction of magnetic torque, it cannot be applied along the geomagnetic induction vector

$\mathbf{M}_{\text{magn}} = \mathbf{m} \times \mathbf{B}_{\text{magn}}$ is the control torque

where

\mathbf{m} is the satellite dipole moment

\mathbf{B}_{magn} is the geomagnetic induction vector

Solution:

construction of an optimal magnetically controllable attitude trajectory using PSO algorithm



<https://www.discovermagazine.com/environment/earths-magnetic-field-probably-isnt-reversing>

Particle swarm optimization (PSO)

the non-gradient biologically inspired global optimization method

Optimization problem:

$$\min_{x \in \mathbb{U}} \Phi(x)$$

$\Phi(x)$ is the cost function

\mathbb{U} is the search space

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Swarm characteristics:

P is the number of particles in the swarm

\mathbf{x}_p is the particle position

v_p is the particle velocity

G is the number of generations

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Position and velocity change:

$$\mathbf{x}_p(i+1) = \mathbf{x}_p(i) + \mathbf{v}_p(i+1)$$

$$\mathbf{v}_p(i+1) = c_{in} \mathbf{v}_p(i) + c_{cog} [\mathbf{x}_{p,best}(i) - \mathbf{x}_p(i)] + c_{soc} [\mathbf{x}_{best}(i) - \mathbf{x}_p(i)]$$

i is the current generation number

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*The inertial component
it is responsible for the search
continuation in the same direction*

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*The cognitive component
the desire to return to its own better
position found earlier*

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*The social component
representing striving for a better
position found in the particle vicinity*

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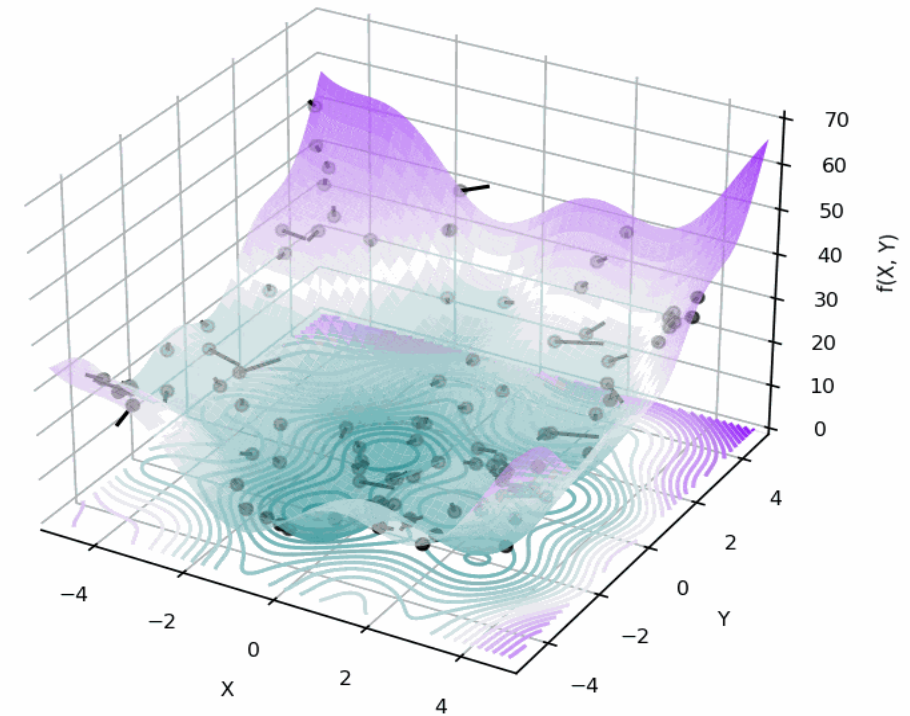
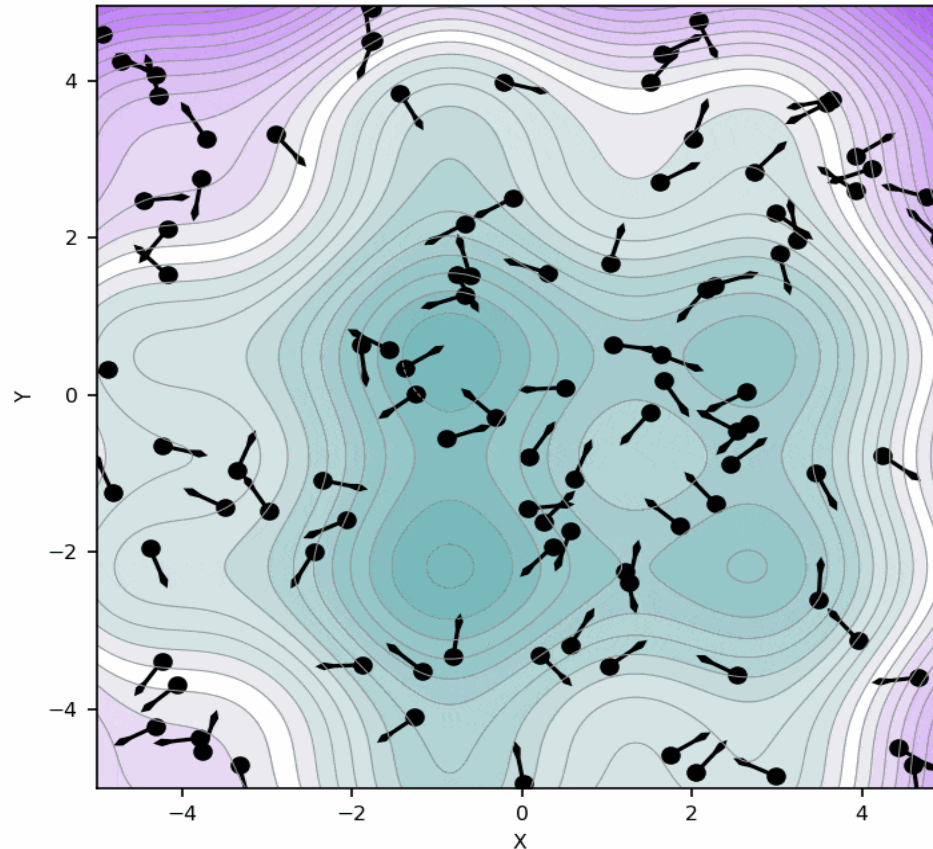
Search stop criteria:

- 1) the cost function derivative is small (cost function stagnation)
- 2) all particles are falling into some neighborhood of the best position (swarm stagnation)

Example

$$[1/100] \quad c_{in} = 0.8, \quad c_{cog} = 2.0, \quad c_{soc} = 2.0$$

<https://towardsdatascience.com/particle-swarm-optimization-visually-explained-46289eeb2e14>



$$\Phi(x, y) = x^2 + (y + 1)^2 - 5\cos(1.5x + 1.5) - 3\cos(2x - 1.5) \rightarrow \min$$
$$\mathbb{U} = \{x, y \mid -5 \leq x \leq 5, -5 \leq y \leq 5\}$$

Motion model

Orbit parameters:

$h = 550$ km – attitude

$i = 57^\circ$ – inclination

$T_o \approx 1.58$ h – orbital period

Satellite parameters:

$10 \times 20 \times 30$ cm – shape (parallelepiped)

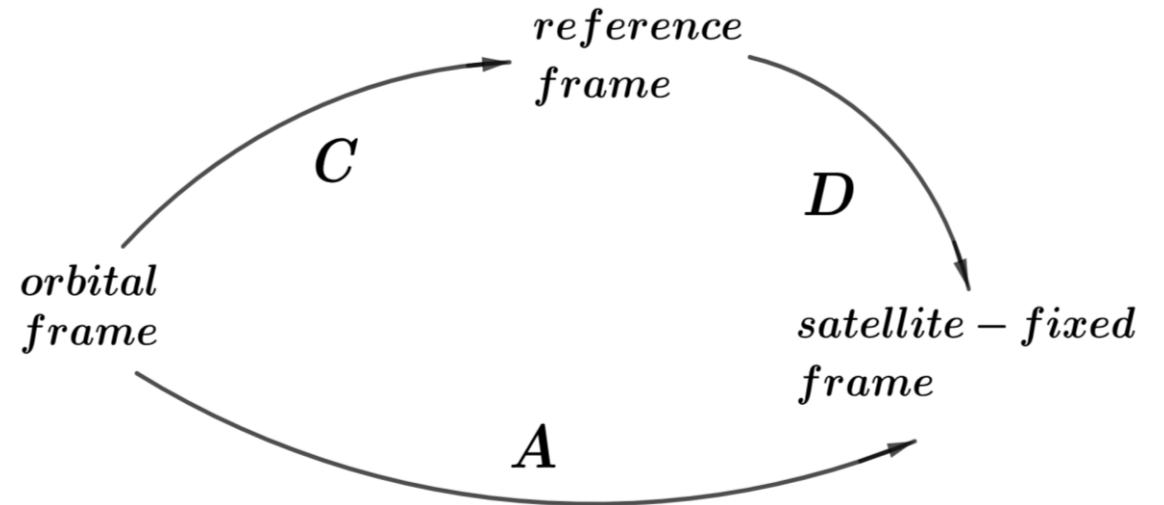
$c = (0, 1, 0)$ cm – center of mass displacement

$J = \text{diag}(0.15, 0.13, 0.11)$ kg·m² – inertia tensor

Equations of motion:

$$\begin{cases} \dot{\omega}_{abs} = J^{-1} (M_{ctrl} + M_{grav} + M_{aero} + \\ \quad + M_{dist} - \omega_{abs} \times J \omega_{abs}) \\ \dot{q}_o = -0.5 q^T \omega, \\ \dot{q} = 0.5 (q_o \omega + q \times \omega) \end{cases}$$

Reference frames:



Trajectory construction

Stage 1

search for a trajectory on which the projection of the control torque onto the Earth's geomagnetic induction vector is minimal

Attitude trajectory:

$$\alpha(t) = a_1 \sin \omega_0 t + a_2 \cos \omega_0 t + a_3 \sin 2\omega_0 t + a_4 \cos 2\omega_0 t,$$

$$\beta(t) = b_1 \sin \omega_0 t + b_2 \cos \omega_0 t + b_3 \sin 2\omega_0 t + b_4 \cos 2\omega_0 t,$$

$$\gamma(t) = g_1 \sin \omega_0 t + g_2 \cos \omega_0 t + g_3 \sin 2\omega_0 t + g_4 \cos 2\omega_0 t,$$

ω_0 – orbital angular velocity

Cost function:

$$\Phi_1 = \frac{dt}{T_0} \sqrt{\sum_{t_0=0}^{T_0} \left(\frac{\mathbf{M}_{ctrl}(t)}{|\mathbf{M}_{ctrl}(t)|}, \frac{\mathbf{B}_{magn}(t)}{|\mathbf{B}_{magn}(t)|} \right)^2} \rightarrow \min$$

$$\mathbb{U} = \{a_k, b_k, g_k \in (-3.5 \cdot 10^{-2}, 3.5 \cdot 10^{-2}) \text{ rad}, k = \overline{1,4}\}$$

Required control torque:

$$\mathbf{M}_{ctrl}(\alpha, \beta, \gamma) = \mathbf{J} \dot{\boldsymbol{\omega}}_{abs}(\alpha, \beta, \gamma) + \boldsymbol{\omega}_{abs}(\alpha, \beta, \gamma) \times \mathbf{J} \boldsymbol{\omega}_{abs}(\alpha, \beta, \gamma) - \mathbf{M}_{grav}(\alpha, \beta, \gamma) - \mathbf{M}_{aero}(\alpha, \beta, \gamma)$$

Trajectory construction

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search for a trajectory on which the projection of the control torque onto the Earth's geomagnetic induction vector is minimal

Attitude trajectory parameters finding by PSO:

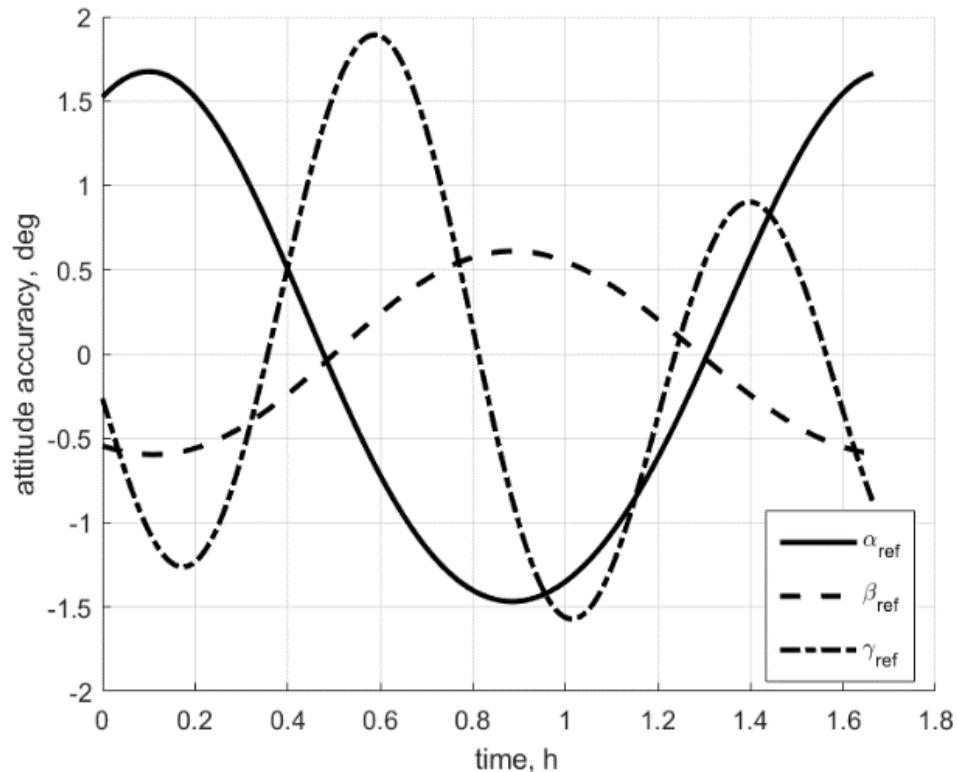
$$\begin{aligned} a_1 &= 1.016 \cdot 10^{-2} \text{ rad}, & b_1 &= -4.028 \cdot 10^{-3} \text{ rad}, & g_1 &= 8.067 \cdot 10^{-3} \text{ rad}, \\ a_2 &= 2.545 \cdot 10^{-2} \text{ rad}, & b_2 &= -9.717 \cdot 10^{-3} \text{ rad}, & g_2 &= -4.127 \cdot 10^{-3} \text{ rad}, \\ a_3 &= 1.449 \cdot 10^{-3} \text{ rad}, & b_3 &= -4.841 \cdot 10^{-5} \text{ rad}, & g_3 &= -2.433 \cdot 10^{-2} \text{ rad}, \\ a_4 &= 1.124 \cdot 10^{-3} \text{ rad}, & b_4 &= 2.231 \cdot 10^{-4} \text{ rad}, & g_4 &= -4.637 \cdot 10^{-4} \text{ rad}. \end{aligned}$$

Trajectory construction

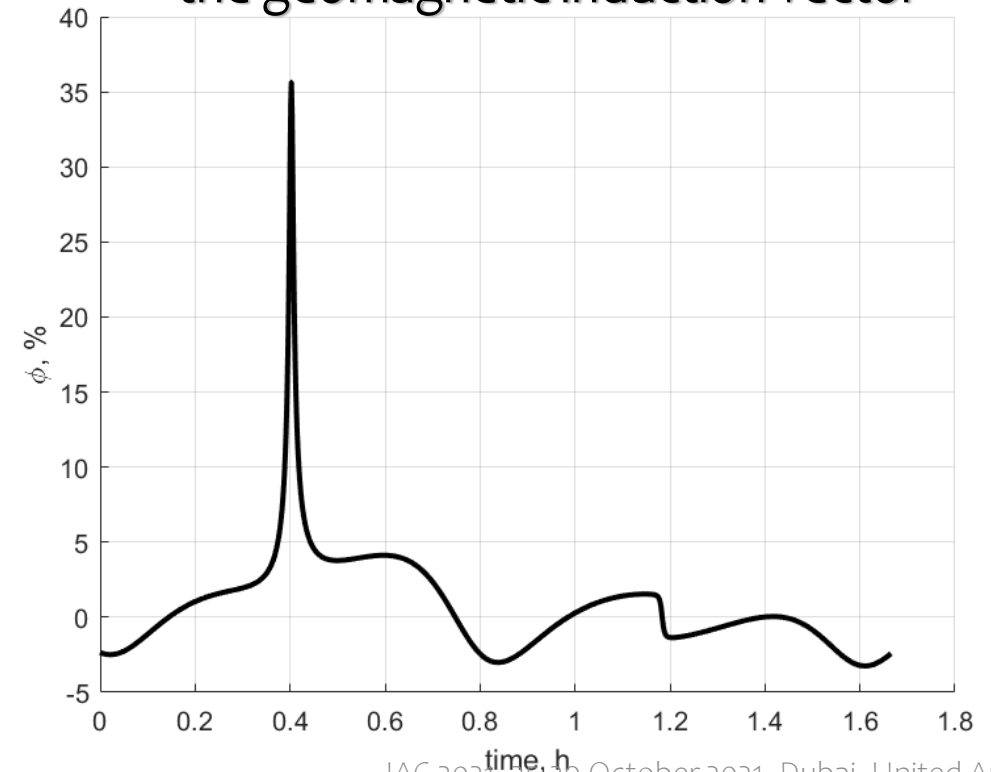
Stage 1

search for a trajectory on which the projection of the control torque onto the Earth's geomagnetic induction vector is minimal

Reference trajectory



Control torque projection onto the geomagnetic induction vector



Trajectory construction

Stage 2

construct a magnetic control that provides convergence

Required control torque (using Lyapunov function):

$$M_{ctrl} = -k_{\omega} \omega_{rel} - k_a S + \omega_{abs} \times J \omega_{abs} + J \dot{A}(\omega_0 + \omega_{ref}) + J A \dot{\omega}_{ref} - M_{grav} - M_{aero}$$

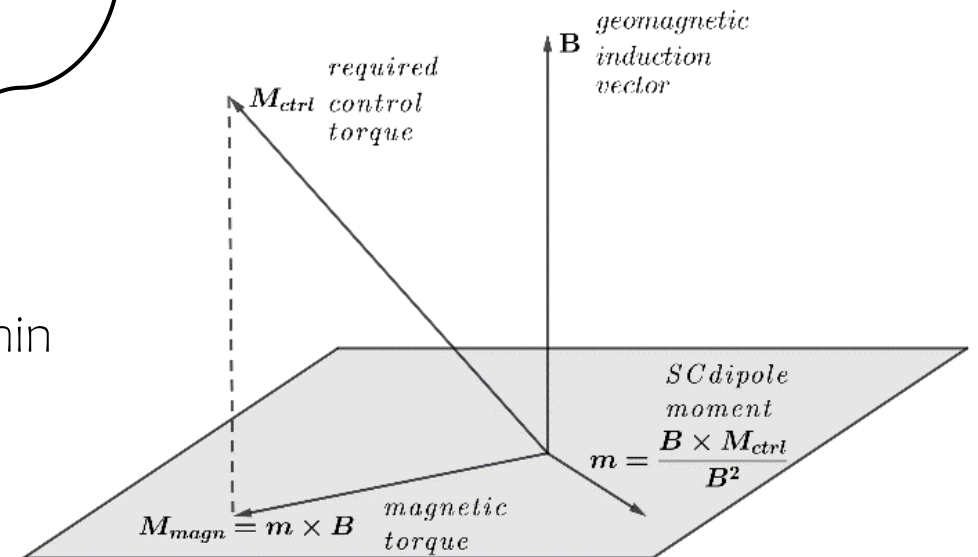
Receiving magnetic control torque:

$$M_{magn} = m \times B_{magn} = \frac{B_{magn} \times M_{ctrl} \times B_{magn}}{B_{magn}^2}$$

Cost function:

$$\Phi_2 = \left(\sum_{t_0=0}^{T_0} \left((\omega_{rel}^T B_{magn}) \cdot (M_{ctrl}^T B_{magn}) \right)^2 + \sum_{t_0=0}^{T_0} \left((S^T B_{magn}) \cdot (M_{ctrl}^T B_{magn}) \right)^2 \right) \rightarrow \min$$

$$\mathbb{U} = \{k_{\omega} \in (5 \cdot 10^{-5}, 10^{-2}) \text{ Nms}, k_a \in (10^{-8}, 5 \cdot 10^{-5}) \text{ Nm}\}$$



Trajectory construction

Stage 2

construct a magnetic control that provides convergence

Control gains finding by PSO:

$$k_{\omega} = 1.119 \cdot 10^{-4} \text{ Nms}, k_a = 6.578 \cdot 10^{-8} \text{ Nm}$$

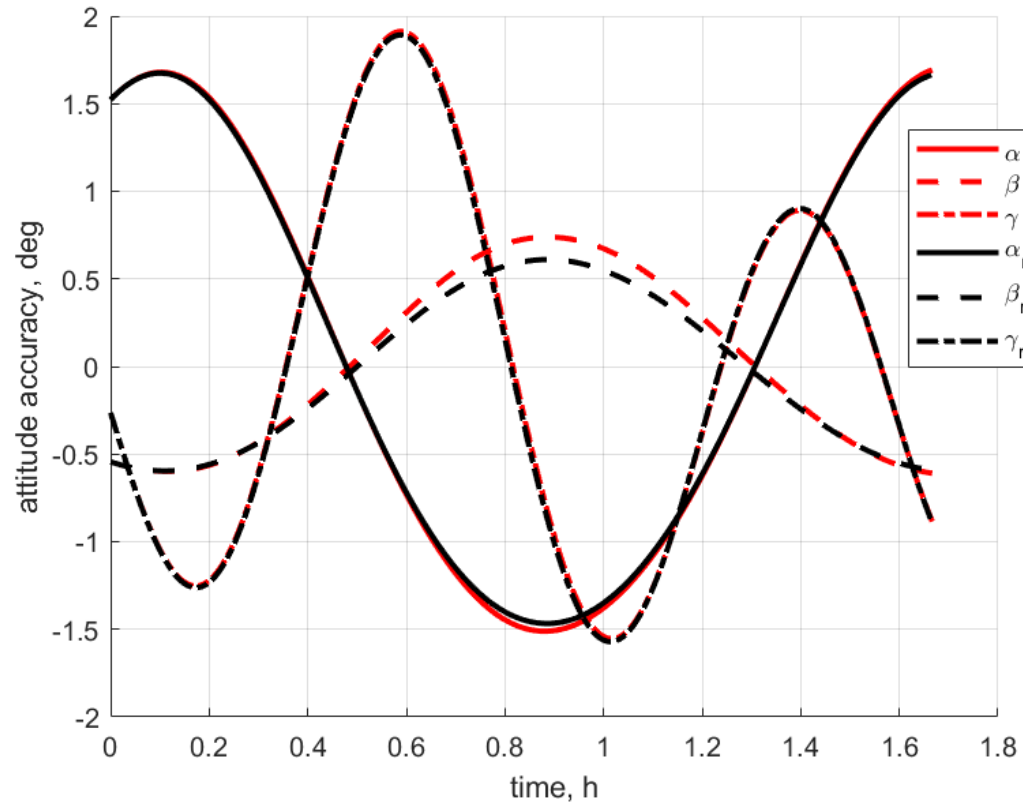
* To prove the **asymptotic stability** of the resulting motion, we linearize the equations of motion and use the **Floquet theory**. The norm of eigenvalues of the obtained monodromy matrix (**characteristic multipliers, ρ**) lie inside the unit circle: $|\rho| < 1$

Trajectory construction

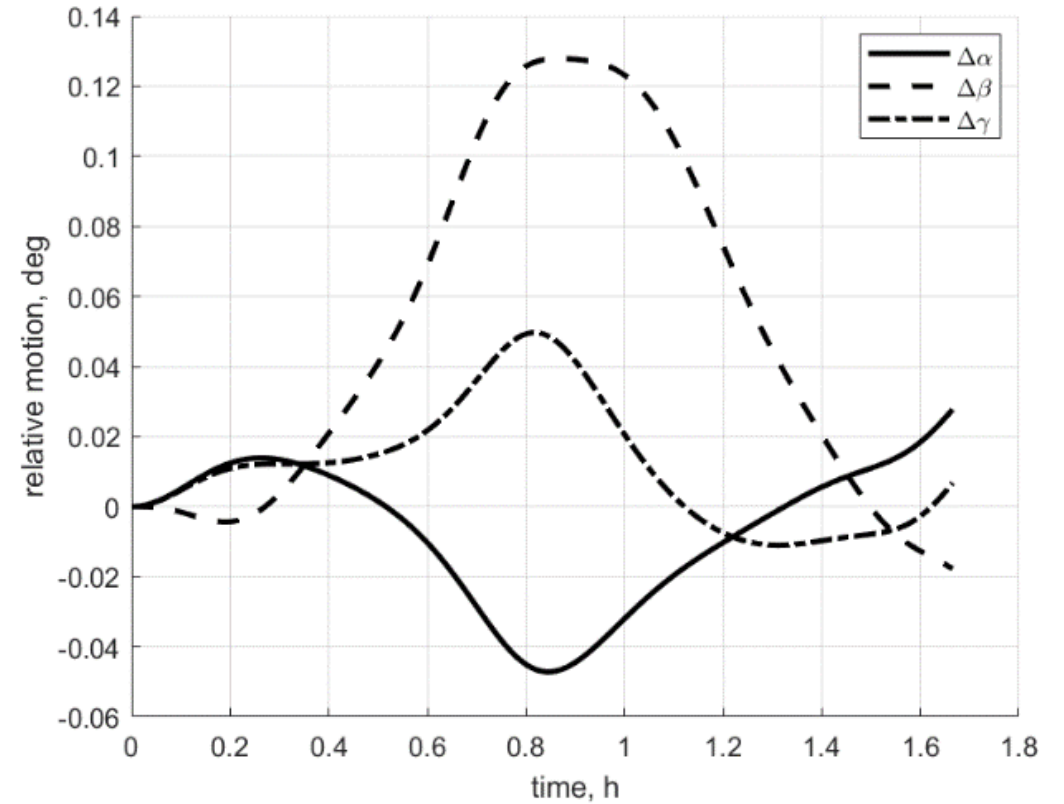
Stage 2

construct a magnetic control that provides convergence

Reference and real trajectory



Reference trajectory deviation

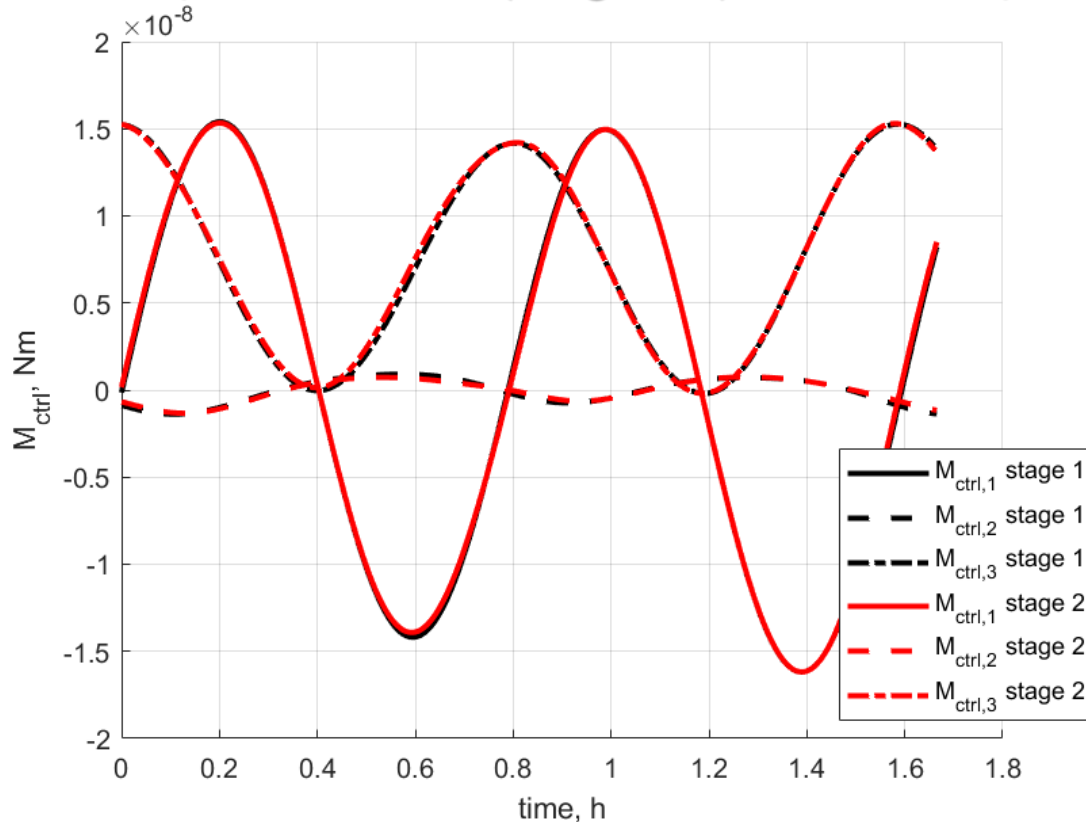


Trajectory construction

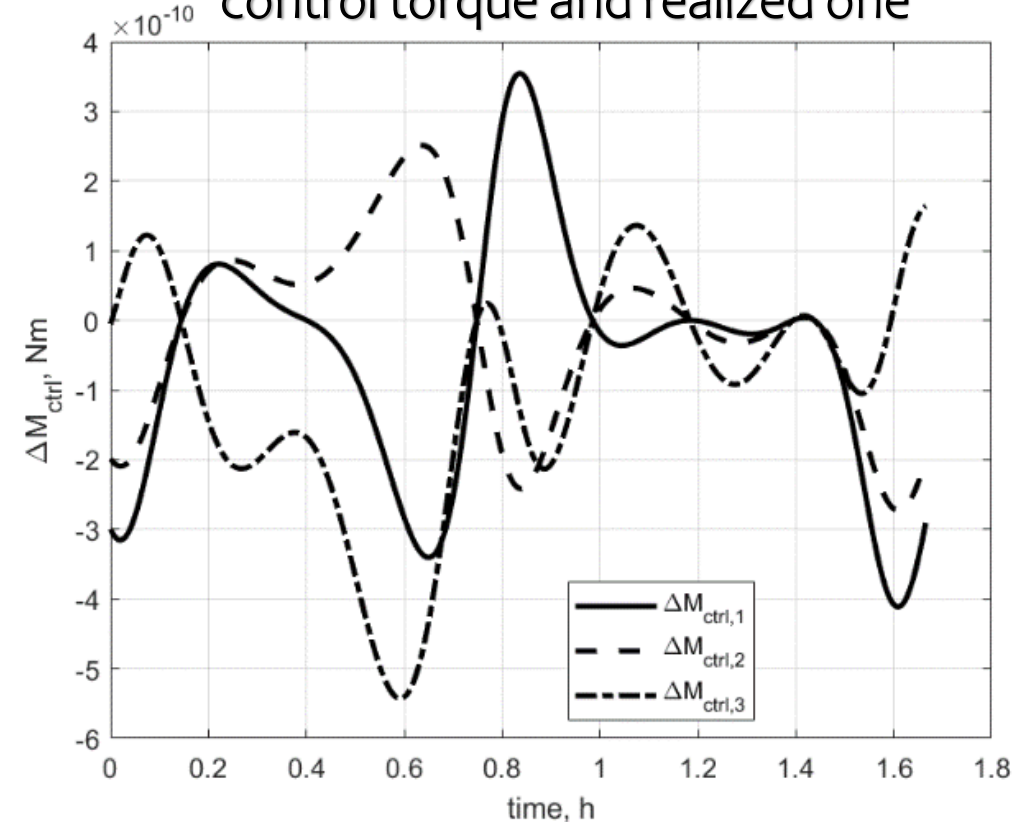
Stage 2

construct a magnetic control that provides convergence

Required and real (magnetic) control torque



The difference between the required control torque and realized one



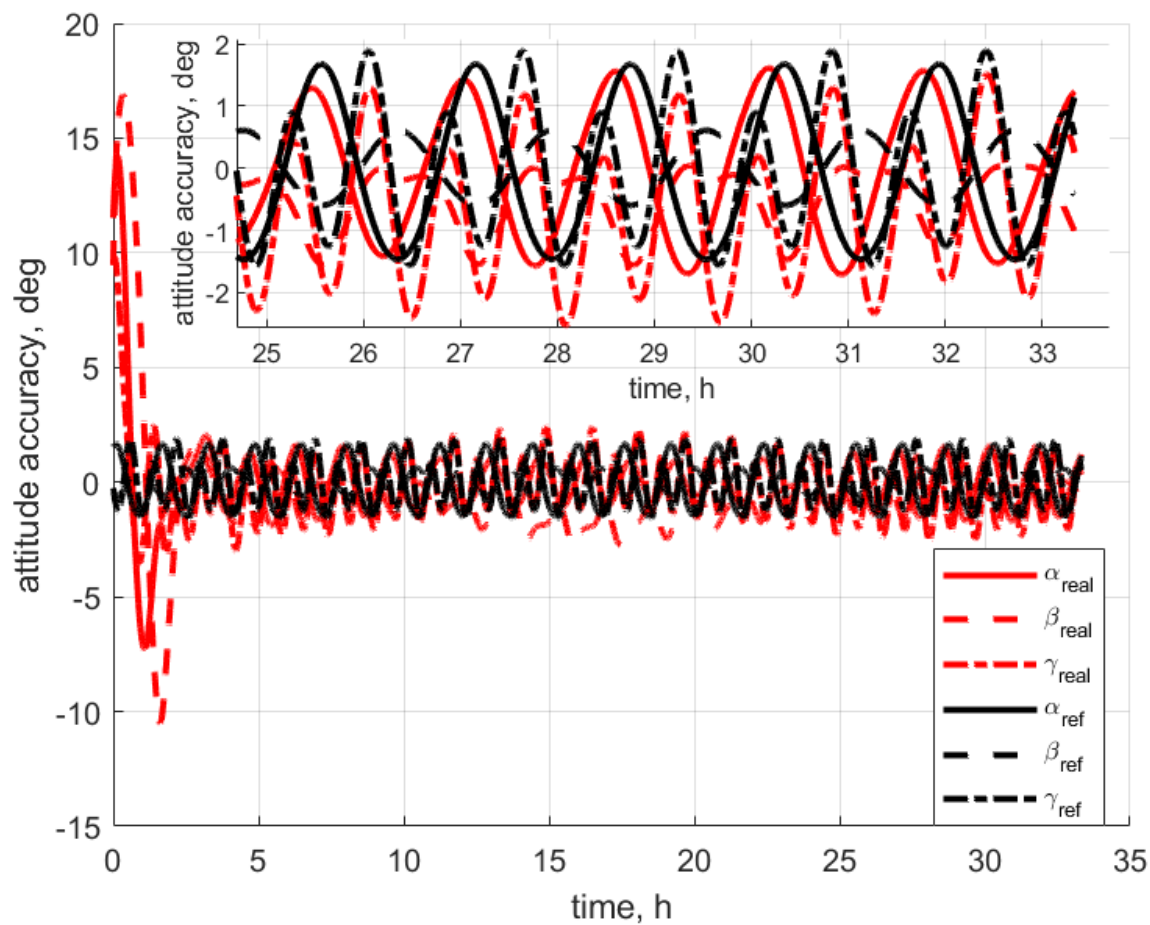
Numerical example

Parameters for numerical simulation

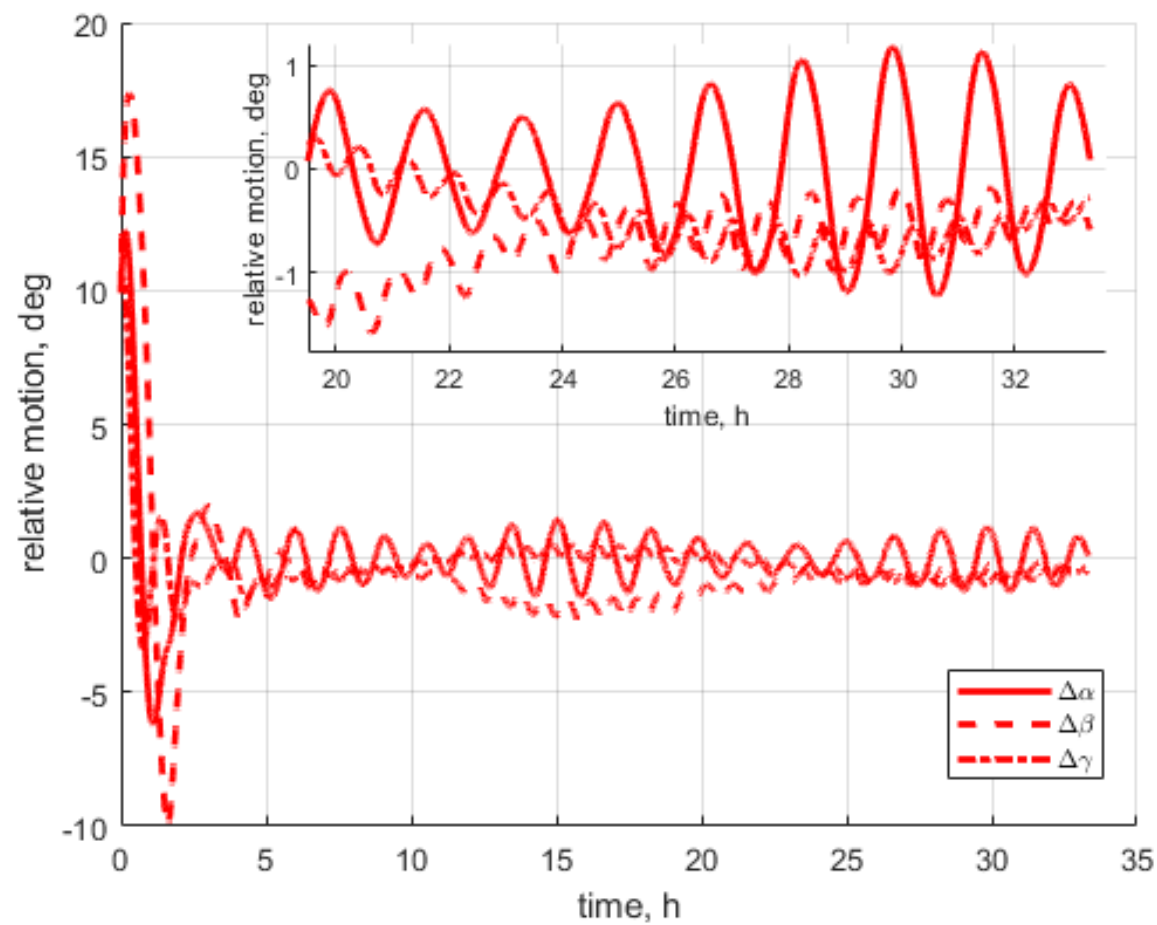
Simulation time	$T = 20T_0 \approx 33 \text{ h}$
SC initial angular velocity	$\omega = [1, 2, 3] \cdot 10^{-4} \text{ rad/s}$
SC initial orientation	$\alpha = 11.5^\circ \cdot \frac{\pi}{180^\circ} \approx 0.2 \text{ rad}$ $\beta = 9.5^\circ \cdot \frac{\pi}{180^\circ} \approx 0.165 \text{ rad}$ $\gamma = 9.7^\circ \cdot \frac{\pi}{180^\circ} \approx 0.17 \text{ rad}$
Magnetic field model	inclined dipole
Inaccuracy of knowledge of the density of the atmosphere	20%
External random disturbances	$M_{dist} \sim 10^{-9} \text{ Nm}$

Numerical example

Reference and real trajectory

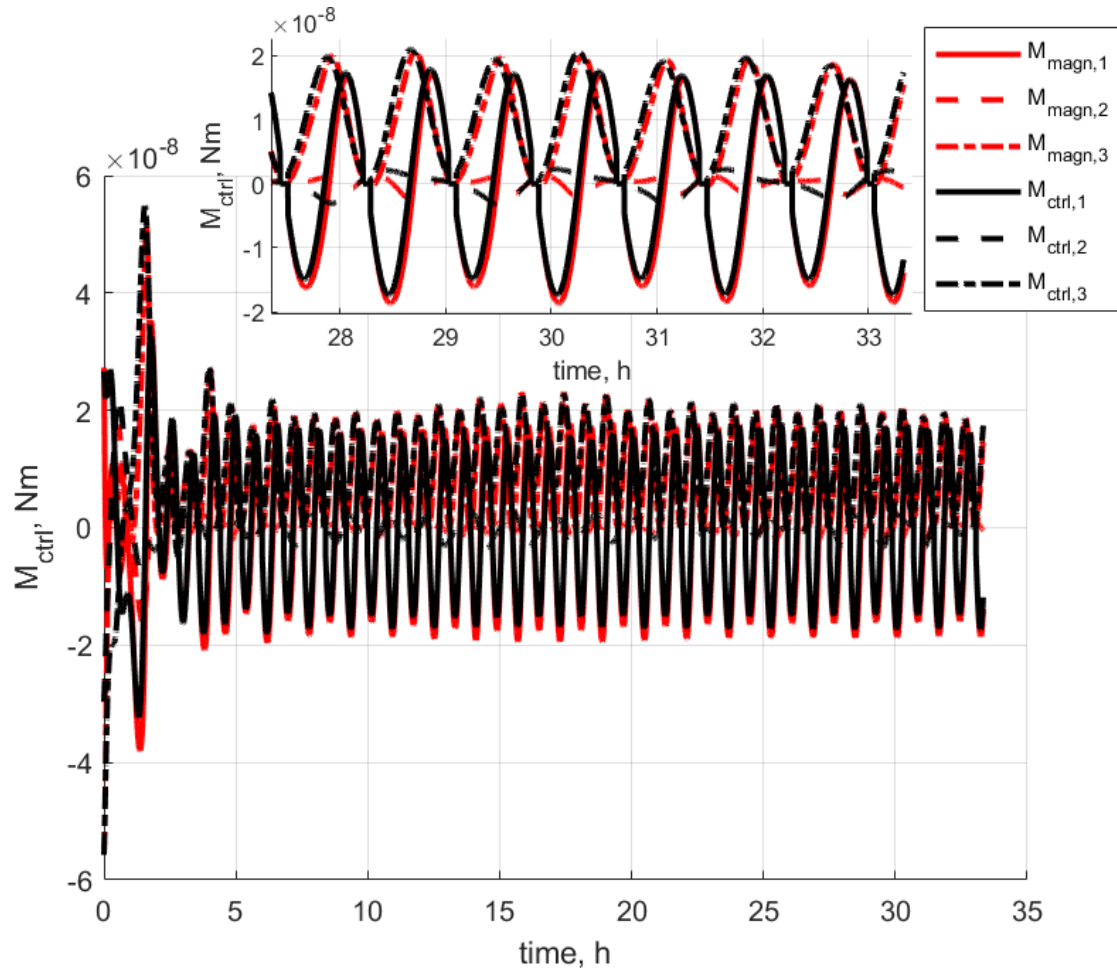


Reference trajectory deviation

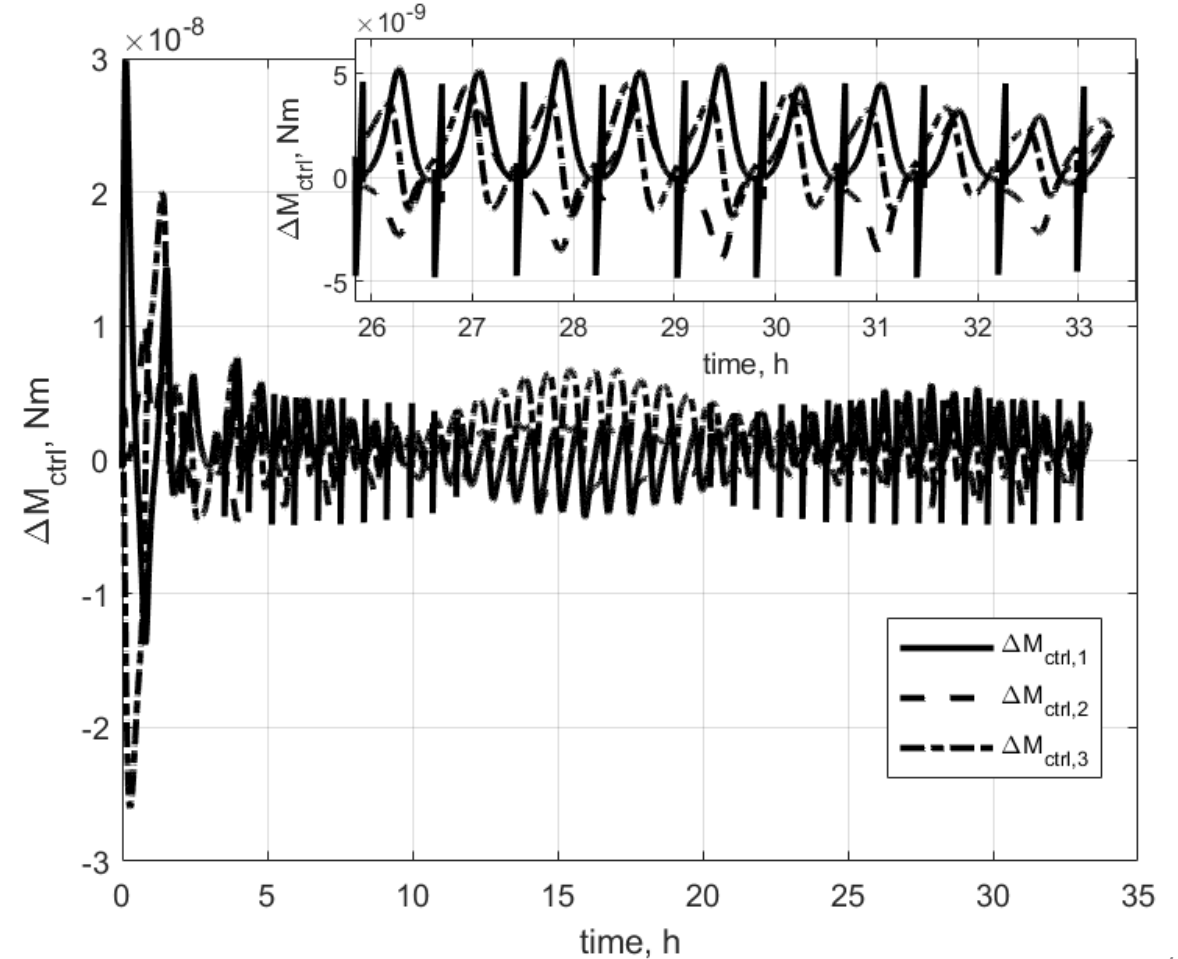


Numerical example

Required and real (magnetic) control torque



The difference between the required control torque and realized one



Conclusion

- a **method for constructing an attitude trajectory** is proposed
- the **particle swarm optimization (PSO)** method is applied to find the **optimal trajectory coefficients and optimal control gains**
- the magnetic attitude control system fully maintains the trajectory
- **numerical example** is given
- the orientation **accuracy** is about **2 degrees**

Thank you for listening!